

Variational Methods for Latent Variable Problems

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Outline for today:

- Some examples of latent variable models
- A template: The Neyman-Scott “paradox” and marginalization
- Bayesian versus frequentist approaches to marginalization
- The classical EM algorithm (in brief)

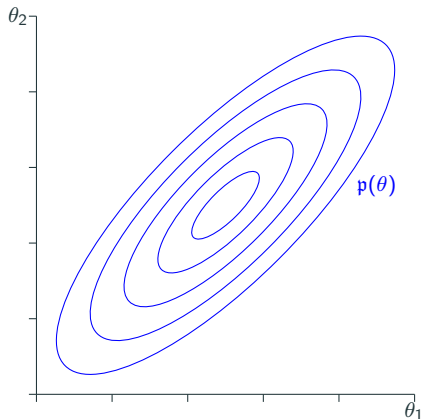
Next week, we will build on these ideas to present more general variational inference.

KL divergence exercises

$p(\theta)$ = Correlated bivariate normal

$\mathcal{Q} = \{\text{All bivariate normals}\}$

What is $q^*(\theta) = \underset{q \in \mathcal{Q}}{\operatorname{argmin}} \operatorname{KL}(q(\theta) || p(\theta))$?

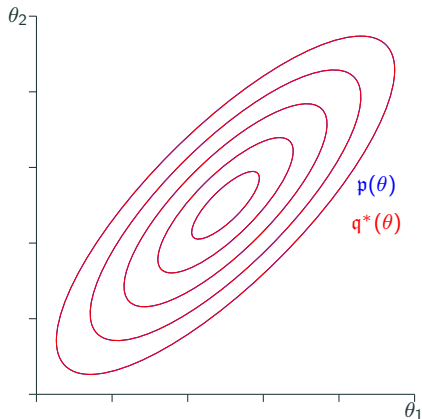


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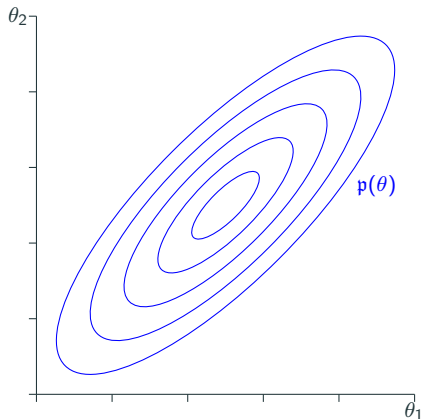


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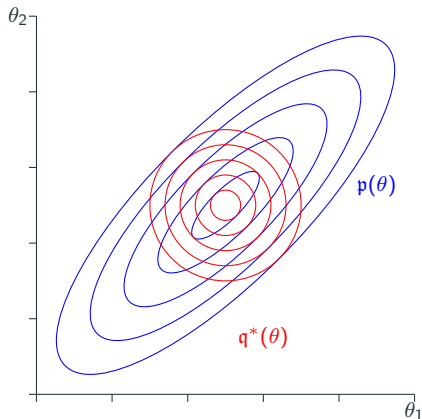


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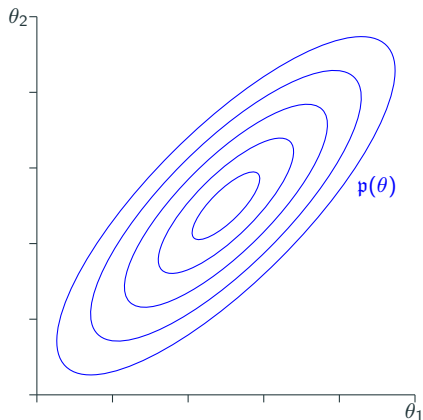


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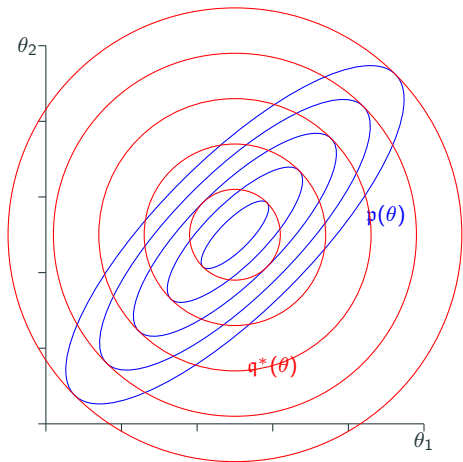


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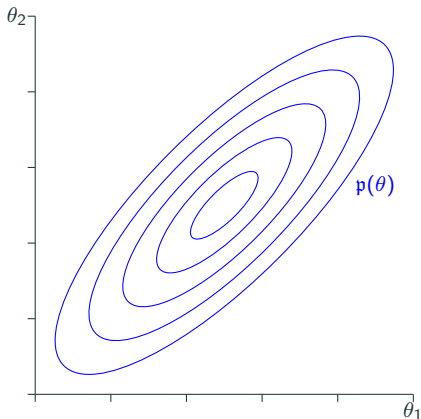
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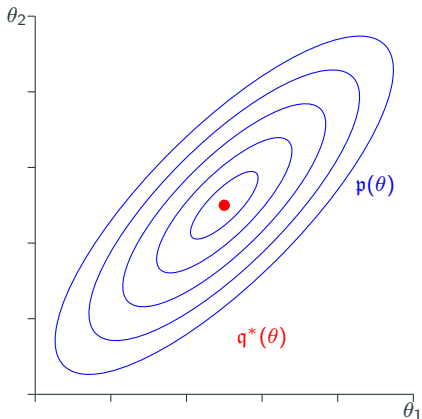
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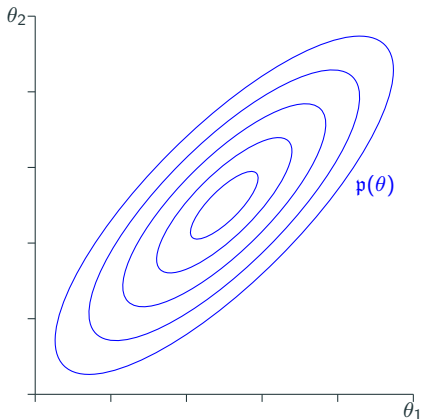
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